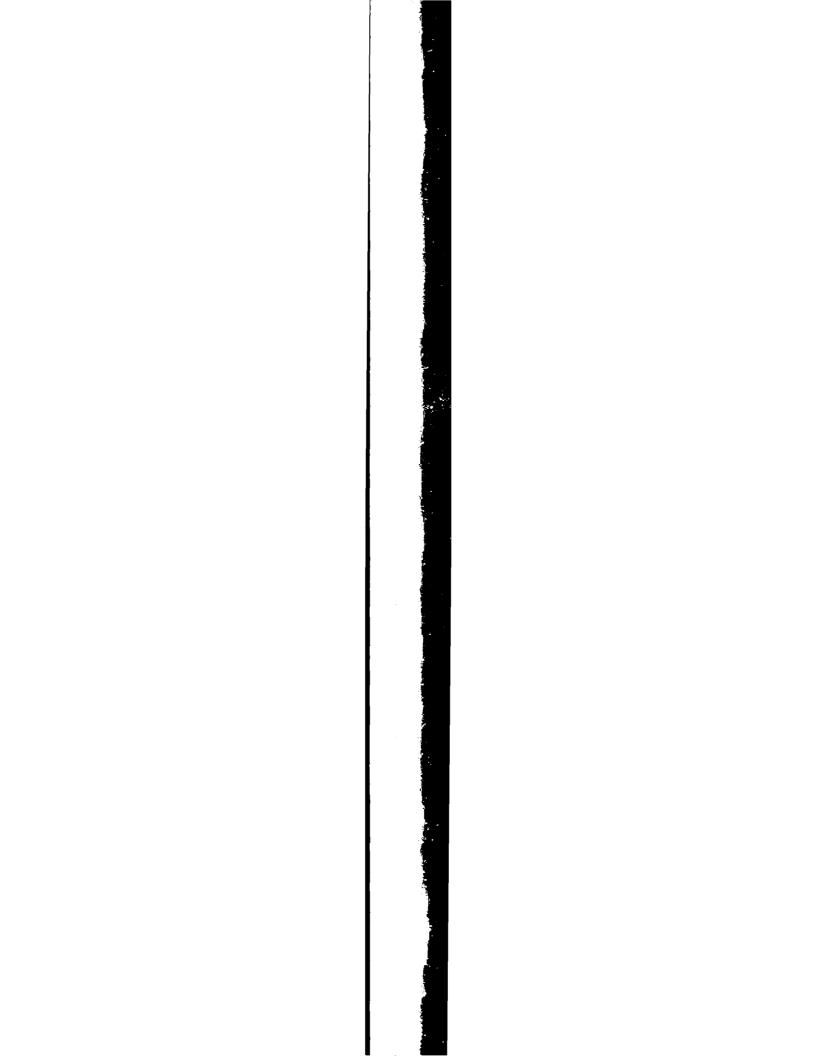
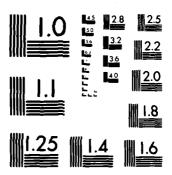
A NOTE ABOUT THE STRONG CONVERGENCE OF THE NONPARAMETRIC ESTIMATION OF A. (U) PITTSBURGH UNIV PA CENTER FOR MULTIVARIATE ANALYSIS Z FANG SEP 84 12/1 TR-84-45 AFOSR-TR-85-8085 F49628-82-K-8081 F/G 12/1 AD-A150 325 1/1 UNCLASSIFIED NL





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UNCLASSIFIED SECURITY CLASSIFICATION OF THIS PAGE REPORT DOCUMENTATION PAGE B REPORT SECURITY CLASSIFICATION 16 RESTRICTIVE MARKINGS UNCLASSIFIED 3 DISTRIBUTION AVAILABILITY OF REPORT 28 SECURITY CLASSIFICATION AUTHORITY Approved for public release; distribution 20 DECLASSIFICATION, DOWNGRADING SCHEDULE unlimited. 5. MONITORING ORGANIZATION REPORT NUMBER(S) A PERFORMING ORGANIZATION REPORT NUMBERIS. AFOSR-TR- 85-0005 TR-84-45 68 NAME OF PERFORMING ORGANIZATION 78 NAME OF MONITORING ORGANIZATION 60. OFFICE SYMBOL University of Pittsburgh elf applicable Air Force Office of Scientific Research 7b. ADDRESS (City, State and ZIP Code 6c ADDRESS (City, State and ZIP Code) Center for Multivariate Analysis Directorate of Mathematical & Information Pittsburgh PA 15260 Sciences, Bolling AFB DC 20332-6448 & NAME OF FUNDING/SPONSORING Bb. OFFICE SYMBOL 9. PROCUREMENT INSTRUMENT IDENTIFICATION NUMBER ORGANIZATION (If applicable) F49620-82-K-0001 **AFOSR** Be ADDRESS (City, State and ZIP Code) 10 SOURCE OF FUNDING NOS WORK UNIT PROGRAM PROJECT TASK ELEMENT NO NO NO NO Bolling AFB DC 20332-6448 A5 61102F 2304 11. TITLE (Include Security Classification) A NOTE ABOUT THE STRONG CONVERGENCE OF THE NONPARAMET RIC ESTIMATION OF A REGRESSION FUNCTION 12. PERSONAL AUTHOR(S) Zhaoben Fang* 134 TYPE OF REPORT 13b. TIME COVERED 14. DATE OF REPORT IYE, Mo., Days 15 PAGE COUNT SEP 84 FROM Technical 16 SUPPLEMENTARY NOTATION *On leave of absence from China University of Science and Technology. 18. SUBJECT TERMS (Continue on reverse if necessary and identify by block number)
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Approved for problem it.

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China University of Science and Technology and Center for Multivariate Analysis
University of Pittsburgh

September 1984

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ABSTRACT

Consider the regression model $Y_i^* = g(x_i^*) + e_i^*$ i = 1, 2...n. X_i^* 's are unordered design variables, g unknown function defined on [0,1]. $\{e_i^*\}$ i.i.d r.v with mean 0 and finite moment of order p > 1. The asymptotic behavior of estimator g_n are studied.

Key Words: Nonparametric regression, Kernel estimation, large sample property.

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We consider the regression model

$$Y_i^* = g(x_i^*) + e_i^*$$
 $i = 1, 2,...,n$ (1)

where $x_i^* \in [0,1]$ are unordered design variables. e_i^* are iid random variables with zero mean. g(x) is an unknown function defined on the interval $\{0,1\}$. The problem is to estimate g(x) through observable variables Y_i^* .

Denote ordered design variables by $x_i \le x_2 \le ... \le x_n$, and define $Y_i = Y_j^*$ and $e_i = e_j^*$ if $x_i = x_j^*$ and $\delta_n = \max_{1 \le i \le n+1} (x_i - x_{i-1})$ where $X_0 = 0$, $X_{n+1} \ge 1$. An estimator proposed by Lin and Cheng [1] is given by

$$g_n(x) = \sum_{i=1}^n Y_i \int_{x_{i-1}}^{x_i} a_n^{-1} k \left(\frac{x-z}{a_n} \right) dz$$
 (2)

where $\{a_n\}$ is a sequence of positive constants converging to 0, as $n + \infty$, k(z) is a kernel density satisfying:

- (i) $k(z) \geq 0$;
- (ii) k(z) = 0 for $z \notin [-LL]$ for some positive constant L;

(iii)
$$\int_{-1}^{L} k(z)dz = 1.$$

In fact, it is obtained by the natural estimator

$$h(x) = \begin{cases} Y_1 & x \le x_1 \\ Y_{i+1} & x_i < x \le x_{i+1} & i = 1,...,n-1 \\ Y_n & x > x_n \end{cases}$$
 (3)

after smoothing with weight function $a_n^{-1}k[(x-z)/a_n]$. Lin and Cheng [1] discussed the strong convergence of g_n and the rates of uniform convergence under a condition

E[e^*] $p < \infty$, $p \ge 2$ among other conditions. In this note we relax the restrictions on the moments of e_i^* and improve the results of [1] and [2]. The results are given in the following theorem:

Theorem Assume that k(z) is bounded, E| $e^{+|p|} < \infty$ for p > 1, $\delta_n = 0(n^{-1})$, $a_n = n^{-d}$, for $0 < d < 1 - \frac{1}{p}$. If g(x) is continuous in [0,1], then for all x ϵ (0.1)

$$g_n(x) + g(x)$$
 a.s. (4)

Proof: Denote $e_i = e_i I(|e_i| \le i^{1/p})$.

$$\begin{split} \tilde{g}_n &= \sum_{i=1}^{11} \tilde{e}_i \int_{x_{i-1}}^{x_i} a_n^{-1} k \left(\frac{x-z}{s_n} \right) \, dz \,, \, \text{where I(') is an indicator function.} \quad \text{Then} \\ g_n(x) &= g(x) - g(x) \\ &= g_n(x) - \tilde{g}_n(x) + \tilde{g}_n(x) - E\tilde{g}_n(x) + E\tilde{g}_n(x) - Eg_n(x) + Eg_n(x) - g(x) \\ &\stackrel{\Delta}{=} I_1 + I_2 + I_3 + I_4. \end{split}$$

In the following we prove $l_i \stackrel{a.s.}{\rightarrow} 0$, i = 1, 2 and $l_j \rightarrow 0$, j = 3, 4 separately. First, when n is large enough

$$I_{4} = Eg_{n} - g = \sum_{i=1}^{n} g(x_{i}) \int_{x_{i-1}}^{x_{i}} a_{n}^{-1} k\left(\frac{x-z}{a_{n}}\right) dz - g(x)$$

$$= \sum_{i=1}^{n} \int_{x_{i-1}}^{x_{i}} [g(x_{i}) - g(z)] a_{n}^{-1} k\left(\frac{x-z}{a_{n}}\right) dz$$

$$+ \sum_{i=1}^{n} \int_{x_{i-1}}^{x_{i}} [g(z) - g(x)] a_{n}^{-1} k\left(\frac{x-z}{a_{n}}\right) dz$$
(6)

By the uniform continuity of g(x) in [0.1] and the property of k(z), the first term above is less than $\varepsilon \int_0^1 \frac{1}{a_n} \, k \, \left(\frac{x-z}{a_n} \right) \, dz$, where ε is arbitrary positive constant. The second term in (6) is not greater than

$$\sup_{|z-x|<\delta} |g(z)-g(x)| + \frac{1}{\delta} \sup_{|t|>\frac{\delta}{a_n}} |tk(t)| \int_0^1 |g(z)| dz$$

$$+ \sup_{|x|} |g(x)| \int_{|u|>\frac{\delta}{a_n}} k(u) du.$$

So it is easy to see when $n \rightarrow \infty$

$$I_A + 0 \tag{7}$$

by Holder and Chebyshev inequality.

$$\begin{aligned} | \, I_{3} \, | &\leq \sum_{i=1}^{n} \mathsf{E} \{ | \, \mathbf{e}_{i} \, | \, | \, | \, (| \, \mathbf{e}_{i} \, | \, > i^{1/p}) \} \, \int_{\mathsf{X}_{i-1}}^{\mathsf{X}_{i}} \mathsf{a}_{n}^{-1} \mathsf{k} \, \left(\frac{\mathsf{x} - \mathsf{z}}{\mathsf{a}_{n}} \right) \, \mathsf{d} \mathsf{z} \\ &\leq \mathsf{a}_{n}^{-1} \, \delta_{n} \, \| \, \mathsf{k} \, \| \, \sum_{i=1}^{n} \, [\mathsf{P}(| \, \mathbf{e}_{i} \, | \, > i^{1/p})]^{1 - \frac{1}{p}} (\mathsf{E}| \, \mathbf{e}_{i} \, |^{p})^{\frac{1}{p}} \\ &\leq \mathsf{a}_{n}^{-1} \, \delta_{n} \, \| \, \mathsf{k} \, \| \, \sum_{i=1}^{n} \, \frac{1}{i \, 1 - \frac{1}{p}} \, (\mathsf{E}| \, \mathbf{e}_{i} |^{p})^{1 - \frac{1}{p}} (\mathsf{E}| \, \mathbf{e}_{i} \, |^{p})^{\frac{1}{p}} \\ &\leq \mathsf{c} \, \mathsf{a}_{n}^{-1} \, \delta_{n} \, \sum_{i=1}^{n} \, \frac{1}{i \, 1 - \frac{1}{p}} \leq \mathsf{c} \, \mathsf{a}_{n}^{-1} \, \delta_{n} \, \mathsf{n}^{\frac{1}{p}} + \mathsf{0}, \qquad \mathsf{n} + \mathsf{\infty} \end{aligned} \tag{8}$$

where $\| k \| = \sup_X | k(x) |$, c a constant. Notice that if $1 < b \le 2$, we have $e^Z \le 1 + z + |z|^b$ when $z \le 1$, so if $r.v Z_i \le 1$, $EZ_i = 0$ then $E(exp\{Z_i\}) \le exp(E||Z_i||^b)$. Now, take b < p, $d_n = \ell_n n \ell_n \ell_n n$, $Z_i = d_n (\tilde{e}_i - E\tilde{e}_i) \int_{X_{i-1}}^{X_i} a_n^{-1} k \left(\frac{x-z}{a_n} \right) dz$. By the choice of d_n and the conditions on δ_n and a_n in the theorem we know that $Z_i \le 1$ when n is large enough. Thus, we have

$$\begin{split} & E\left(\prod_{i=1}^{n} \exp\{Z_{i}\}\right) = \prod_{i=1}^{n} E(\exp\{Z_{i}\}) \leq \prod_{i=1}^{n} \exp(E|Z_{i}|^{b}) \\ & \leq \prod_{i=1}^{n} \exp\{d_{n}(d_{n} \parallel k \parallel \delta_{n}a_{n}^{-1})^{b-1} E|e_{i}|^{b} \int_{X_{i-1}}^{X_{i}} a_{n}^{-1} k \left(\frac{x-z}{a_{n}}\right) dz \} \\ & \leq \exp\{c \ d_{n}(d_{n} \delta_{n}a_{n}^{-1})^{b-1}\} \end{split}$$

For given $\varepsilon > 0$, when n is large enough,

$$P(I_2 > \varepsilon) \le e^{-d_n \varepsilon} E(\exp\{d_n I_2\})$$

$$= e^{-d_n \varepsilon} E(\prod_{i=1}^n \exp\{Z_i\}) < e^{-\frac{1}{2}d_n \varepsilon}.$$

So, we have

$$\sum_{n} P(l_2 > \varepsilon) < \sum_{n} e^{\frac{1}{2}d_n \varepsilon} < \infty$$

by Borel-Cantelli lemma

$$\lim_{n\to\infty}\sup_{n\to\infty}I_2\leq0\qquad \text{a.s.}$$

Similarly, it is true that

$$\lim_{n\to\infty}\inf_{0} l_2 \ge 0 \qquad a.s$$

Therefore

$$l_2 + 0$$
 a.s. (9)

Since E| e_i | p < ∞ , we know that $\sum\limits_{n}^{\infty} P(|e_i| > j^{\overline{p}}) < \infty$. By Borel-Cantelli lemma we have $P(|e_i| > i^{\overline{p}}, i.o) = 0$. Hence

$$\sum_{i=1}^{\infty} e_i^2 i(|e_i| > i^{\overline{p}}) < \infty \text{ a.s.}$$
 (10)

Finally,

$$||\mathbf{I}_1||^2 = |\sum_{i=1}^{n} (\mathbf{e}_i - \mathbf{e}_i)| \int_{\mathbf{x}_{i-1}}^{\mathbf{x}_{i}} a_n^{-1} k \left(\frac{\mathbf{x} - \mathbf{z}}{a_n} \right) d\mathbf{z}|^2$$

By Schwartz inequality,

$$||i_1||^2 \le \sum_{i=1}^n (\int_{x_{i-1}}^{x_i} a_n^{-1}) k(\frac{x-z}{a_n}) dz)^2 \sum_{i=1}^n e_i^2 l(||e_i|| > i^{\overline{p}})$$

$$\leq a_n^{-1} \delta_n \| k \| \int_0^1 a_n^{-1} k \left(\frac{x - z}{a_n} \right) dz \sum_{i=1}^{\infty} e_i^2 I(\| e_i \| > i^{\overline{D}})$$

$$\leq c a_n^{-1} \delta_n \sum_i e_i^2 I(\| e_i \| > i^{\overline{D}}).$$

By (10) we have

$$||1_1|^2 + 0$$
 a.s. (11)

From (5), (7), (8), (9), and (11), the theorem follows.

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